- ¹ Title: Tree height and leaf drought tolerance traits shape growth responses across droughts in a temperate
- 2 broadleaf forest

12

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Results	697		
Discussion	1467		
Acknowledgements	125		

22 Summary

- As climate change is driving increased drought frequency and severity in many forested regions around
 the world, mechanistic understanding of the factors conferring drought resistance in trees is
 increasingly important. The dendrochronological record provides a window through which we can
 understand how tree size and species' traits shape tree growth responses during droughts.
- We analyzed tree-ring records for twelve species that comprise 97% of the woody productivity of the 25.6-ha ForestGEO plot in a broadleaf deciduous forest of northern Virginia (USA) to test hypotheses on how tree height, microenvironment characteristics, and species' traits shaped drought responses across the three strongest regional droughts over a 60-year period (1950 2009).
- Individual-level drought resistance decreased with tree height, which was strongly correlated with
 exposure to higher evaporative demand and solar radiation. The potentially greater rooting volume of
 larger trees did not confer an advantage in sites with low topographic wetness index. Resistance was
 greater among species whose leaves experienced less shrinkage upon desiccation and lost turgor (wilted)
 at more negative water potentials.
- We conclude that tree height and leaf drought tolerance traits influence growth responses during
 drought, as recorded in the tree-ring record spanning historical droughts. Thus, these factors can be
 useful for predicting future drought responses under climate change.
- Key words: annual growth; crown exposure; drought; Forest Global Earth Observatory (ForestGEO); leaf drought tolerance traits; temperate broadleaf deciduous forest; tree height; tree-ring

1 Introduction

Forests play a critical global role in climate regulation (Bonan, 2008), yet there remains enormous uncertainty as to how the forest-dominated terrestrial carbon sink will respond to climate change (Friedlingstein et al., 2006). An important aspect of this uncertainty lies with physiological responses of trees 44 to drought (Kennedy et al., 2019). In many forested regions around the world, the risk of severe drought is 45 increasing (Trenberth et al., 2014; Dai et al., 2018), often despite increasing precipitation (Intergovernmental Panel on Climate Change, 2015; Cook et al., 2015). Droughts, intensified by climate change, have been affecting forests worldwide and are expected to continue as one of the most important drivers of forest change in the future (Allen et al., 2010, 2015). Understanding forest responses to drought requires elucidation of how tree size, microenvironment, and species' traits jointly influence individual-level drought resistance, and the extent to which their influence is consistent across droughts. Because the resistance and 51 resilience of growth to drought is linked to trees' probability of surviving drought (DeSoto et al., 2020; Liu et al., 2019), understanding growth responses can also help elucidate which trees are most vulnerable to drought-induced mortality. However, it has proven difficult to resolve the many factors affecting tree growth during drought with available forest census data, which only rarely captures extreme drought, and with tree-ring records, which capture multiple droughts but usually only sample a subset of a forest community, typically focusing on a single species or the largest individuals. 57 Many studies have shown that within and across species, large trees tend to be more affected by drought. Greater growth reductions for larger trees were first shown on a global scale by Bennett et al. (2015), and subsequent studies have reinforced this finding (e.g., Hacket-Pain et al., 2016; Gillerot et al., 2020). It has yet to be resolved which of several potential underlying mechanisms most strongly shape these trends in drought 61 response. First, tree height itself may be a primary driver. Taller trees face the biophysical challenge of lifting water greater distances against the effects of gravity and friction (McDowell et al., 2011; McDowell 63 and Allen, 2015; Ryan et al., 2006; Couvreur et al., 2018). Vertical gradients in stem and leaf traits-including smaller and thicker leaves (higher leaf mass per area, LMA), greater resistance to hydraulic dysfunction (i.e., more negative water potential at 50% loss of hydraulic conductivity, more negative P50), and lower hydraulic conductivity at greater heights (Couvreur et al., 2018; Koike et al., 2001; McDowell et al., 2011)—enable trees 67 to become tall (Couvreur et al., 2018). Greater stem capacitance (i.e., water storage capacity) of larger trees may also confer resistance to transient droughts [Phillips et al. (2003)*; Scholz et al. (2011)]. Indeed, tall trees require xylem of greater hydraulic efficiency, such that xylem conduit diameters are wider in the basal portions of taller trees, both within and across species (Olson et al., 2018; Liu et al., 2019), and throughout the conductive systems of angiosperms (Zach et al., 2010; Olson et al., 2014, 2018). Wider xylem conduits 72 plausibly make large trees more vulnerable to embolism during drought (Olson et al., 2018), and traits 73 conducive to efficient water transport may also lead to poor ability to recover from or re-route water around embolisms (Roskilly et al., 2019). Larger trees may also have lower drought resistance because of microenvironmental and ecological factors. Their crowns tend to occupy more exposed canopy positions, which are associated with higher evaporative demand (Kunert et al., 2017). Subcanopy trees tend to fare better specifically due to the benefits of a buffered environment (Pretzsch et al., 2018). Counteracting the liabilities associated with tall height, large 79 trees tend to have larger root systems (Enquist and Niklas, 2002), potentially mitigating some of the biophysical challenges they face by allowing greater access to water. Larger root systems-if they grant access to deeper water sources—would be particularly advantageous in drier microenvironments (e.g., hilltops, as

compared to valleys and streambeds) during drought. Finally, tree size-related responses to drought can be modified by species' traits and their distribution across size classes (Meakem et al., 2018; Liu et al., 2019). Understanding the mechanisms driving the greater relative growth reductions of larger trees during drought requires sorting out the interactive effects of height and associated exposure, root water access, and species' traits. Debates have also arisen regarding the traits influencing tree growth responses to drought. Studies within temperate broadleaf forests have observed ring-porous species showing higher drought tolerance than diffuse-porous species (Friedrichs et al., 2009; Elliott et al., 2015; Kannenberg et al., 2019), but this distinction would not hold in the global context (Wheeler et al., 2007; Olson et al., 2020) and does not resolve differences among the many species within each category. Commonly-measured traits including wood 92 density and leaf mass per area (LMA) have been linked to drought responses within some temperate 93 deciduous forests (Abrams, 1990; Guerfel et al., 2009; Hoffmann et al., 2011; Martin-Benito and Pederson, 2015) and across forests worldwide (Greenwood et al., 2017). However, in other cases these traits could not explain drought tolerance (e.g., in a tropical rainforest; Maréchaux et al., 2019), or the direction of response was not always consistent. For instance, higher wood density has been associated with greater drought 97 resistance at a global scale (Greenwood et al., 2017), but correlated negatively with tree performance during

In contrast, hydraulic traits have direct physiological linkages to tree growth and mortality responses to 102 drought. For instance, water potentials at which percent the loss of conductivity surpasses a certain 103 threshold (e.g., P50 and P88, representing 50 and 88% loss of conductivity, respectively) and hydraulic safety 104 margin (i.e., difference between typical minimum water potentials and P50 or P88) correlate with drought 105 performance across global forests (Anderegg et al., 2016). However, these are time-consuming to measure and 106 therefore infeasible for predicting or modeling drought responses in highly diverse forests (e.g., in the tropics). 107 More easily-measurable leaf drought tolerance traits that have direct linkage to plant hydraulic function can 108 explain variation in plant distribution and function (Medeiros et al., 2019). These include leaf area shrinkage 109 upon desiccation (PLA_{dry} ; Scoffoni et al., 2014) and the leaf water potential at turgor loss point (π_{tlp}), i.e., 110 the water potential at which leaf wilting occurs (Bartlett et al., 2016a; Zhu et al., 2018). Both traits correlate 111 with hydraulic vulnerability and drought tolerance as part of unified plant hydraulic systems (Scoffoni et al., 2014; Bartlett et al., 2016a; Zhu et al., 2018; Farrell et al., 2017). The abilities of both PLA_{dry} and π_{tlp} to 113 explain tree drought resistance remains untested. 114

drought in a broadleaf deciduous forest in the southeastern United States (Hoffmann et al., 2011). Thus, the perceived influence of these traits on drought resistance may actually reflect indirect correlations with other

traits that more directly drive drought responses (Hoffmann et al., 2011).

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Here, we examine how tree height, microenvironment characteristics, and species' traits collectively shape 115 drought resistance, defined as the ratio of annual growth in a drought year to that which would be expected in the absence of drought based on previous years' growth. We test a series of hypotheses and associated 117 specific predictions (Table 1) based on the combination of tree-ring records from the three strongest droughts 118 over a 60-year period (1950 - 2009), species trait measurements, and census and microenvironmental data from a large forest dynamics plot in Virginia, USA. First, we focus on how tree size, alone and in its 120 interaction with microenvironmental gradients, influences drought resistance. We examine the contemporary 121 relationship between tree height and microenvironment, including growing season meteorological conditions and crown exposure. We then test whether, consistent with most forests globally, larger-diameter, taller trees 123 tend to have lower drought resistance in this forest, which is in a region (eastern North America) represented

by only two studies in the global review of Bennett et al. (2015). We also test for an influence of potential access to available soil water, which should be greater for larger trees in dry but not in perpetually wet microsites. Finally, we focus on the role of species' traits, testing the hypothesis that species' traits—particularly leaf leaf drought tolerance traits—predict drought resistance. We test predictions that drought resistance is higher in ring-porous than semi-ring and diffuse-porous species and that it is correlated with wood density—either positively (Greenwood et al., 2017) or negatively (Hoffmann et al., 2011) and positively correlated with LMA. We further test predictions that species with low PLA_{dry} have higher drought resistance, and that species whose leaves lose turgor lower water potentials (more negative π_{tlp}) have higher resistance.

134 Materials and Methods

135 Study site and microclimate

Research was conducted at the 25.6-ha ForestGEO (Forest Global Earth Observatory) study plot at the
Smithsonian Conservation Biology Institute (SCBI) in Virginia, USA (38°53'36.6"N, 78°08'43.4"W; Fig. S1)
(Bourg et al., 2013; Anderson-Teixeira et al., 2015a). SCBI is located in the central Appalachian Mountains
near the northern boundary of Shenandoah National Park. Elevations range from 273 to 338 m above sea
level with a topographic relief of 65m (Bourg et al., 2013). Climate is humid temperate, with mean annual
temperature of 12.7°C and precipitation of 1005 mm yr⁻¹ during our study period (1960-2009; source: CRU
TS v.4.01; Harris et al., 2014). Dominant tree taxa within this secondary forest include *Liriodendron*tulipifera, oaks (Quercus spp.), and hickories (Carya spp.; Table 2).

144 Identifying drought years

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We identified the three largest droughts within the time period 1950-2009, defining drought (Slette et al., 2019) as events with anomalously dry peak growing season climatic conditions. Specifically, we used the metric of Palmer Drought Severity Index (PDSI) during May-August (MJJA; Table S1), which were identified by Helcoski et al. (2019) as the months of the current year to which annual tree growth was most sensitive at this site. PDSI divisional data for Northern Virginia were obtained from NOAA (https://www7.ncdc.noaa.gov/CDO/CDODivisionalSelect.jsp) in December 2017. Based on this, we identified the three strongest droughts during the study period (Figs. 1, S1; Table S1).

was preceded by two years of moderate drought during the growing season and severe to extreme drought starting the previous fall. In August 1966, *PDSI* reached its lowest monthly value (-4.82) of the three droughts. The 1977 drought was the least intense throughout the growing season, and it was preceded by 2.5 years of near-normal conditions, making it the mildest of the three droughts. The 1999 drought was preceded by wetter than average conditions until the previous June, but *PDSI* plummeted below -3.0 in October 1998 and remained below this threshold through August 1999.

The droughts differed in intensity and antecedent moisture conditions (Fig. S1, Table S1). The 1966 drought

Data collection and preparation

Within or just outside the ForestGEO plot, we collected data on a suite of variables including tree heights, microenvironment characteristics, and species traits (Table 3). The SCBI ForestGEO plot was censused in 2008, 2013, and 2018 following standard ForestGEO protocols, whereby all free-standing woody stems ≥ 1cm diameter at breast height (DBH) were mapped, tagged, measured at DBH, and identified to species (Condit,

1998). From these census data, we used measurements of DBH from 2008 to calculate historical DBH and data for all stems ≥ 10 cm to analyze functional trait composition relative to tree height (all analyses described below). Census data are available through the ForestGEO data portal (www.forestgeo.si.edu). 166 We analyzed tree-ring data (xylem growth increment) from 571 trees representing the twelve dominant 167 species (Table 2; Fig. S2). Selected species were those with the greatest contributions to woody aboveground 168 net primary productivity $(ANPP_{stem})$ and together comprised 97% of study plot $ANPP_{stem}$ between 2008 and 2013 (Helcoski et al., 2019). Cores (one per tree) were collected within the ForestGEO plot at breast 170 height (1.3m) in 2010-2011 or 2016-2017. In 2010-2011, cores were collected from randomly selected live trees 171 of each species that had at least 30 individuals \geq 10 cm DBH (Bourg et al., 2013). Annual tree mortality censuses were initiated in 2014 (Gonzalez-Akre et al., 2016), and in 2016-2017, cores were collected from all 173 trees found to have died since the previous year's census. We note that drought was probably not a cause of 174 mortality for these trees, as monthly May-Aug PDSI did not drop below -1.75 in these years or the three years prior (2013-2017), and that trees cored dead displayed similar climate sensitivity to trees cored live 176 (Helcoski et al., 2019). Cores were sanded, measured, and crossdated using standard procedures, as detailed 177 in (Helcoski et al., 2019). The resulting chronologies (Fig. 1a) were published in Zenodo (DOI: 10.5281/zenodo.2649302) in association with Helcoski et al. (2019). 179

For each cored tree, we combined tree-ring records and allometric equations of bark thickness to reconstruct DBH for the years 1950-2009. Prior DBH was estimated using the following equation:

$$DBH_Y = DBH_{2008} - 2 * \left[r_{bark,2008} - r_{bark,Y} + \sum_{year=Y}^{2008} r_{ring,Y} \right]$$

Here, Y denotes the year of interest, r_{ring} denotes ring width derived from cores, and r_{bark} denotes bark thickness. Bark thickness was estimated from species-specific allometries based on the bark thickness data

from the site (Anderson-Teixeira et al., 2015b). Specifically, we used linear regression on log-transformed 184 data to relate r_{bark} to diameter inside bark from 2008 data (Table S2), which were then used to determine r_{bark} in the DBH reconstruction. Tree heights (H) were measured by several researchers for a variety of purposes between 2012 and 2019 187 (n=1,518 trees). Methods included direct measurements using a collapsible measurement rod on small trees 188 (NEON, 2018) or a tape measure on recently fallen trees (this study); geometric calculations using clinometer and tape measure (Stovall et al., 2018a) or digital rangefinders (Anderson-Teixeira et al., 2015b; NEON, 2018); and ground-based LiDAR (Stovall et al., 2018b). Rangefinders used either the tangent method 191 (Impulse 200LR, TruPulse 360R) or the sine method (Nikon ForestryPro) for calculating heights. Both methods are associated with some error (Larjavaara and Muller-Landau, 2013), but in this instance there was 193 no clear advantage of one or the other. Measurements from the National Ecological Observatory Network 194 (NEON) were collected near the ForestGEO plot following standard NEON protocol, whereby vegetation of short stature was measured with a collapsible measurement rod, and taller trees with a rangefinder (NEON, 2018). Species-specific height allometries were developed (Table S3) using log-log regression 197 $(\ln[H] \sim \ln[DBH])$. For species with insufficient height data to create reliable species-specific allometries (n=2, JUNI and FRAM), heights were calculated from an equation developed by combining the height 199 measurements across all species. We then used these allometries to estimate H for each drought year, Y, 200 based on reconstructed DBH_Y . The distribution of H across drought years is shown in Fig. S3.

To characterize how environmental conditions vary with height, data were obtained from the NEON tower located <1km from the study area via the neonUtilities package (Lunch et al., 2020). We used wind speed, 203 relative humidity, and air temperature data, all measured over a vertical profile spanning heights from 7.2 m 204 to above the top of the tree canopy (31.0 or 51.8m, depending on censor), for the years 2016-2018 (NEON, 205 2018). After filtering for missing and outlier values, we determined the daily minima and maxima, which we then aggregated at the monthly scale. 207 Crown position—a categorical variable classifying trees based on exposure to sunlight—was recorded for all 208 cored trees that remained standing during the growing season of 2018 following the protocol of Jennings et al. 209 (1999). Trees were classified as follows: dominant trees were defined as those with crowns above the general 210 level of the canopy, co-dominant trees as those with crowns within the the canopy; intermediate trees as 211 those with crowns below the canopy level, but illuminated from above; and suppressed as those below the 212 canopy and receiving minimal direct illumination from above.

Topographic wetness index (TWI), used here as a metric of long-term mean moisture availability, was 214 calculated using the dynatopmodel package in R (Fig. S2) (Metcalfe et al., 2018). Originally developed by 215 Beven and Kirkby (1979), TWI was part of a hydrological run-off model and has since been used for a 216 number of purposes in hydrology and ecology (Sørensen et al., 2006). TWI calculation depends on an input 217 of a digital elevation model (DEM; ~3.7 m resolution from the elevatr package (Hollister, 2018)), and from 218 this yields a quantitative assessment defined by how "wet" an area is, based on areas where run-off is more 219 likely. From our observations in the plot, TWI performed better at categorizing wet areas than the Euclidean 220 distance from the stream. 221

Species' trait data were collected in August 2018 (Tables 2-3; Fig. S4). We sampled small, sun-exposed 222 branches up to eight meters above the ground from three individuals of each species in and around the 223 ForestGEO plot. Sampled branches were re-cut under water at least two nodes above the original cut and 224 re-hydrated overnight in covered buckets under opaque plastic bags before measurements were taken. 225 Rehydrated leaves taken towards the apical end of the branch (n=3 per individual: small, medium, and large) were scanned, weighed, dried at 60° C for ≥ 48 hours, and then re-scanned and weighed. Leaf area 227 was calculated from scanned images using the LeafArea R package (Katabuchi, 2019). LMA was calculated 228 as the ratio of leaf dry mass to fresh area. PLA_{dry} was calculated as the percent loss of area between fresh and dry leaves. Wood density was calculated for ~1cm diameter stem samples (bark and pith removed) as 230 the ratio of dry weight to fresh volume, which was estimated using Archimedes' displacement. We used the 231 rapid determination method of Bartlett et al. (2012) to estimate osmotic potential at turgor loss point (π_{tlp}) . Briefly, two 4 mm diameter leaf discs were cut from each leaf, tightly wrapped in foil, submerged in liquid 233 nitrogen, perforated 10-15 times with a dissection needle, and then measured using a vapour pressure 234 osmometer (VAPRO 5520, Wescor, Logan, UT, USA). Osmotic potential (π_{osm}) given by the osmometer was used to estimate (π_{tlp}) using the equation $\pi_{tlp} = 0.832\pi_{osm}^{-0.631}$ (Bartlett et al., 2012). 236

237 Statistical Analysis

For each drought year, we calculated a metric drought resistance (Rt) as the ratio of basal area increment (BAI; i.e., change in cross-sectional area) during the drought year to the mean BAI over the five years preceding the drought (Lloret et al., 2011). Thus, Rt values <1 and >1 indicate growth reductions and increases, respectively. Because the Rt metric could be biased by directional pre-drought growth trends, we also tried an intervention time series analysis (ARIMA, (Hyndman et al., 2020)) that predicted mean

drought-year growth based on trends over the preceding ten years and used this value in place of the 243 five-year mean in calculations of resistance (Rt_{ARIMA} = observed BAI/ predicted BAI). The two metrics were strongly correlated (Fig. S5). Visual review of the individual tree-ring sequences with the largest 245 discrepancies between these metrics revealed that Rt was less prone to unreasonable estimates than 246 Rt_{ARIMA} , so we selected Rt as our focal metric, presenting parallel results for Rt_{ARIMA} in the Supplementary Info. In this study we focus exclusively on drought resistance metrics (Rt or Rt_{ARIMA}), and 248 not on the resilience metrics described in Lloret et al. (2011), because (1) we would expect resilience to be 249 controlled by a different set of mechanisms, and (2) the findings of (DeSoto et al., 2020) suggest that Rt is a 250 more important drought response metric for angiosperms in that low resistance to moderate droughts was a 251 better predictor of mortality during subsequent severe droughts than the resilience metrics. 252 Analyses focused on testing the predictions presented in Table 1 with Rt as the response variable, and then 253 repeated using Rt_{ARIMA} as the response variable. Models were run for all drought years combined and for each drought year individually. The general statistical model for hypothesis testing was a mixed effects 255 model, implemented in the lme4 package in R (Bates et al., 2019). In the multi-year model, we included a 256 random effect of tree nested within species and a fixed effect of drought year to represent the combined 257 effects of differences in drought characteristics. Individual year models included a random effect of species. 258 All models included fixed effects of independent variables of interest (Tables 1,3) as specified below. All 259 variables across all best models had variance inflation factors <1.2 (1 + /- 0.019). We used AICc to assess model selection, and conditional/marginal R-squared to assess model fit as implemented in the AICcmodavg 261 package in R (Mazerolle and portions of code contributed by Dan Linden., 2019). AICc refers to a corrected 262 version of AICc, and is best suited for small data sizes (see Brewer et al., 2016). To avoid over-fitting models with five species traits (Table 3) across only 12 species, we did not include all 264 traits as fixed effects in a single linear mixed model, but rather conducted individual tests of each species 265 trait to determine the relative importance and appropriateness for inclusion in the main model. These tests 266 followed the model structure specified above, then added ln[H] and ln[TWI] to create a base model against 267 which we tested traits. Trait variables were considered appropriate for inclusion in the main model if they 268 had a consistent direction of response across all droughts and if their addition to the base model improved fit 269 (at $\triangle AICc \ge 1.0$) in at least one drought year (Table S4). We note that we did not use the $\triangle AICc \ge 1.0$ 270 criterion as a test of significance, but rather of whether the variable had enough influence to be considered as 271 a candidate variable in full models. We then determined the top full models for predicting Rt (or Rt_{ARIMA}). To do so, we compared models with all possible combinations of candidate variables, including $ln[H]^*ln[TWI]$ and species traits as specified 274 above. We identified the full set of models within $\triangle AICc=2$ of the best model (that with lowest AICc). 275 When a variable appeared in all of these models and the sign of the coefficient was consistent across models, we viewed this as support for the acceptance/rejection of the associated prediction (Table 1). If the variable 277 appeared in some but not all of these models, and its sign was consistent across models, we considered this 278 partial support/rejection. In presentation of the results below, we note instances where the Rt_{ARIMA} model disagreed with the Rt model, but otherwise do not discuss the Rt_{ARIMA} model. 280 All analysis beyond basic data collection was performed using R version 3.6.2 (R Core Team, 2019). Other 281 R-packages used in analyses are listed in the Supplementary Information (Appendix S1). All data, code, and 282 results are available through the SCBI-ForestGEO organization on GitHub 283 (https://github.com/SCBI-ForestGEO: SCBI-ForestGEO-Data and McGregor_climate-sensitivity-variation repositories), with static versions corresponding to data and analyses presented here archived in Zenodo (DOIs: 10.5281/zenodo.3604993 and [TBD], respectively.

7 Results

- 288 Tree height and microenvironment
- 289 In the years for which we have vertical profiles in climate data (2016-2018), taller trees—or those in dominant
- crown positions—were generally exposed to higher evaporative demand during the peak growing season
- months (May-August; Fig. 2). Specifically, maximum daily wind speeds were significantly higher above the
- top of the canopy (40-50m) than within and below (10-30m) (Fig. 2a). Relative humidity was also somewhat
- lower during June-August, ranging from ~50-80% above the canopy and ~60-90% in the understory (Fig. 2b).
- ²⁹⁴ Air temperature did not vary consistently across the vertical profile (Fig. 2c).
- 295 Crown position varied as expected with height (dominant > co-dominant > intermediate > suppressed), but
- with substantial variation (Fig. 2d). There were significant differences in height across all crown position
- classes (Fig. 2d). A comparison test between height and crown position data from the most recent
- ²⁹⁸ ForestGEO census (2018) revealed a correlation of 0.73.
- 299 Community-level drought responses
- At the community level, cored trees showed substantial growth reductions in all three droughts, with a mean
- Rt of 0.86 in 1966 and 1999, and 0.84 in 1977 (Fig. 2b). Across the entire study period (1950-2009), the
- focal drought years were the three years with the largest fraction of trees exhibiting $Rt \leq 0.7$. Specifically, in
- each drought, roughly 30% of the cored trees had growth reductions of > 30% (Rt < 0.7): 29% in 1966, 32%
- in 1977, and 27% in 1999. However, some individuals exhibited increased growth, i.e., Rt > 1.0: 26% of trees
- 305 in 1966, 22% in 1977, and 26% in 1999.
- In the context of the multivariate model, Rt did not vary across drought years. That is, drought year as a
- variable did not appear in any of the top models -i.e., models that were statistically indistinguishable
- $(\Delta AICc < 2)$ from the best model.
- 309 Tree height, microenvironment, and drought resistance
- Taller trees (based on H in the drought year) showed stronger growth reductions during drought (Table 1;
- Figs. 4, S6). Specifically, ln[H] appeared, with a negative coefficient, in the best model ((Δ AICc=0) and all
- top models when evaluating the three drought years together (Tables S6-S7). The same held true for 1966
- individually. For the 1977 drought, ln[H] did not appear in the best model, but was included, with a
- negative coefficient, among the top models-i.e., models that were statistically indistinguishable ($\Delta AICc < 2$)
- from the best model (Tables 1, S6-S7). For the 1999 drought, ln[H] had no significant effect.
- Rt had a significantly negative response to ln[TWI] across all drought years combined (Figs. 4, S6, Table
- $_{317}$ S6-S7). The effect was also significant for 1977 and 1999 individually (Fig. 4, Table S6). When Rt_{ARIMA}
- was used as the response variable, the effect was significant in 1977, and included in some of the top models
- in 1966 and 1999 (Table S7). This negates the idea that trees in moist microsites would be less affected by
- drought. Nevertheless, we tested for a ln[H] * ln[TWI] interaction, a negative sign of which could indicate
- that smaller trees (presumably with smaller rooting volume) are more susceptible to drought in drier
- microenvironments with a deeper water table. This hypothesis was rejected, as the ln[H] * ln[TWI]
- interaction was never significant, and had a positive sign in any top models in which it appeared (Tables 1,

S6-S7). This term did appear with a positive coefficient in the best Rt_{ARIMA} model for all years combined (Table S7), indicating that the negative effect of height on Rt was significantly stronger in wetter microhabitats. 326 Species' traits and drought resistance 327 Species, as a factor in ANOVA, had significant influence (p< 0.05) on all traits (wood density, LMA, 328 PLA_{dry} , and π_{tlp}), with more significant pairwise differences for wood density and PLA_{dry} than for LMA

and π_{tlp} (Table 2, Fig. S4). Drought resistance also varied across species, overall and in each drought year (Fig. 3). Significant differences in Rt across species were most pronounced in 1966 with a total of seven 331 distinct groupings, while 1977 had four and 1999 had two. Averaged across all droughts, Rt was lowest in 332 Liriodendron tulipifera (mean Rt = 0.66) and highest in Fagus grandifolia (mean Rt = 0.99). Wood density, LMA, and xylem porosity were all poor predictors of Rt (Tables 1,S4-S5). Wood density and 334 LMA were never significantly associated with Rt in the single-variable tests and were therefore excluded 335 from the full models. Xylem porosity was also excluded from the full models, as it had no significant 336 influence for all droughts combined and had contrasting effects in the individual droughts: whereas 337 ring-porous species had higher Rt than diffuse- and semi-ring- porous species in the 1966 and 1999 droughts, 338 they had lower Rt in 1977 (Table S4). It is noteworthy that the two diffuse-porous species in our study, Liriodendron tulipifera and Fagus grandifolia, were at opposite ends of the Rt spectrum (Fig. 3), further 340 refuting the idea that xylem porosity is a useful predictor of Rt in the context of this study. 341 In contrast, PLA_{dry} , and π_{tlp} were both negatively correlated to drought resistance (Figs. 4, S6; Tables 342 1,S4-S7). Both had consistent signs across all droughts, and their inclusion at least marginally improved the

model ($\Delta AICc > 1.0$) for at least one of the three droughts (Table S4), qualifying them as candidate 344 variables for the full model. PLA_{dry} had a significant influence, with negative coefficient, in full models for 345 the three droughts combined and for the 1966 drought individually (Fig. 4; Tables S6-S7). For 1977 and 1999, it was included with a negative coefficient in some of the top models (Tables S6-S7). π_{tlp} was included 347 with a negative coefficient in the best model for both all droughts combined and for the 1977 drought 348 individually (Fig. 4; Table S6). It was also included in some of the top models for 1999 (Tables S6-S7).

Discussion

329

Tree height, microenvironment, and leaf drought tolerance traits shaped tree growth responses across three 351 droughts at our study site (Table 1, Fig. 4). The greater susceptibility of larger trees to drought, similar to 352 forests worldwide (Bennett et al., 2015), was driven primarily by their height (Stovall et al., 2019). Taller 353 height was likely a liability in itself, and was also associated with greater exposure to conditions that would promote water loss and heat damage during drought (Fig. 2). There was no evidence that greater 355 availability of, or access to, soil water availability increased drought resistance; in contrast, trees in wetter 356 topographic positions had lower Rt (Zuleta et al., 2017; Stovall et al., 2019), and the larger potential rooting volume of large trees provided no advantage in the drier microenvironments. The negative effect of height on 358 Rt held after accounting for species' traits, which is consistent with recent work finding height had a stronger 359 influence on mortality risk than forest type during drought (Stovall et al., 2020). Drought resistance was not consistently linked to species' LMA, wood density, or xylem type (ring- vs. diffuse porous), but was 361 negatively correlated with leaf drought tolerance traits (PLA_{dry}, π_{tlp}) . This is the first study to our 362 knowledge linking PLA_{dry} and π_{tlp} to growth reduction during drought. The directions of these responses

were consistent across droughts (Table S6), supporting the premise that they were driven by fundamental 364 physiological mechanisms. However, the strengths of each predictor varied across droughts (Fig. 4; Tables S6-S7), indicating that drought characteristics interact with tree size, microenvironment, and traits to shape 366 which individuals are most affected. These findings advance our knowledge of the factors that make trees 367 vulnerable to growth declines during drought and, by extension, likely make them more vulnerable to mortality (Sapes et al., 2019). 369 The droughts considered here were of a magnitude that has occurred with an average frequency of 370 approximately once every 10-15 years (Fig. 1a, Helcoski et al., 2019) and had substantial but not devastating 371 impacts on tree growth (Figs. 1b). These droughts were classified as severe (PDSI < -3.0; 1977) or extreme (PDSI < -4.0; 1966, 1999) at our site and have been linked to tree mortality in the eastern United States 373 (Druckenbrod et al., 2019). However, extreme, multiannual droughts such as the so-called "megadroughts" of 374 this type that have triggered massive tree die-off in other regions (e.g., Allen et al., 2010; Stovall et al., 2019) have not occurred in the Eastern United States within the past several decades (Clark et al., 2016). Of the 376 droughts considered here, the 1966 drought, which was preceded by two years of dry conditions (Fig. S1), 377 severely stressed a larger portion of trees (Fig. 1b). The tendency for large trees to have lowest resistance 378 was most pronounced in this drought, consistent with other findings that this physiological response increases 379 with drought severity (Bennett et al., 2015; Stovall et al., 2019). Across all three droughts, the majority of 380 trees experienced reduced growth, but a substantial portion had increased growth (Fig. 1b), potentially due 381 to decreased leaf area of competitors during the drought (REF-if we can find one), and consistent with 382 prior observations that smaller trees can exhibit increased growth rates during drought (Bennett et al., 2015). 383 It is likely because of the moderate impact of these droughts, along with other factors influencing tree growth (e.g., stand dynamics), that our best models characterize only a modest amount of variation in Rt: 11-12% 385 for all droughts combined, and 18-25% for each individual drought (Fig. S6; Table S6). 386 Consistent with studies in other forests worldwide (Bennett et al., 2015), taller trees in this forest exhibited 387 lower drought resistance. Mechanistically, this is consistent with, and reinforces, previous findings that 388 biophysical constraints make it impossible for trees to efficiently transport water to great heights and 389 simultaneously maintain strong resistance and resilience to drought-induced embolism (Olson et al., 2018; 390 Couvreur et al., 2018; Roskilly et al., 2019). Taller trees also face dramatically different microenvironments 391 (Fig. 2). They are exposed to higher wind speeds and lower humidity (Fig. 2a-b), resulting in higher 392 evaporative demand. Unlike other temperate forests where modestly cooler understory conditions have been documented (Zellweger et al., 2019), particularly under drier conditions (Davis et al., 2019), we observed no 394 significant variation in air temperatures across the vertical profile (Fig. 2c). More critically for tree 395 physiology, leaf temperatures can become significantly elevated over air temperature under conditions of high 396 solar radiation and low stomatal conductance (Campbell and Norman, 1998; Rey-Sánchez et al., 2016). 397 Under drought, when air temperatures tend to be warmer, direct solar radiation tends to be higher (because 398 of less cloud cover), and less water is available for evaporative cooling of the leaves, trees with sun-exposed crowns may not be able to simultaneously maintain leaf temperatures below damaging extremes and avoid 400 drought-induced embolism. Indeed, previous studies have shown lower drought resistance in more exposed 401 trees (Liu and Muller, 1993; Suarez et al., 2004; Scharnweber et al., 2019). Unfortunately, collinearity 402 between height and crown exposure in this study (Fig. 2d) makes it impossible to confidently partition 403 causality. Additional research comparing drought responses of early successional and mature forest stands, 404 along with short and tall isolated trees, would be valuable for more clearly disentangling the roles of tree

height and crown exposure. 406 Belowground, taller trees would tend to have larger root systems (Enquist and Niklas, 2002), but this does 407 not necessarily imply that they have greater access to or reliance on deep soil-water resources that may be critical during drought. While tree size can correlate with the depth of water extraction (?), the linkage is 409 not consistent. Shorter trees can vary broadly in the depth of water uptake (?), and larger trees may allocate 410 more to abundant shallow roots that are beneficial for taking up water from rainstorms (Meinzer et al., 1999). In any case, the potentially greater access to water did not override the disadvantage conferred by 412 height-and, in fact, greater moisture access in non-drought years (here, higher TWI) appears to make trees 413 more sensitive to drought (Zuleta et al., 2017; Stovall et al., 2019). This may be because moister habitats 414 would tend to support species and individuals with more mesophytic traits (Bartlett et al., 2016b; 415 Mencuccini, 2003; Medeiros et al., 2019), potentially growing to greater heights (e.g., Detto et al. (2013)), 416 and these are then more vulnerable when drought hits. The observed height-sensitivity of Rt, together with the lack of conferred advantage to large stature in drier topographic positions, agrees with the concept that 418 physiological limitations to transpiration under drought shift from soil water availability to the 419 plant-atmosphere interface as forests age (Bretfeld et al., 2018), such that tall, dominant trees are the most 420 sensitive in mature forests. Again, additional research comparing drought responses across forests with 421 different tree heights and water availability would be valuable for disentangling the relative importance of 422 above- and belowground mechanisms across trees of different size. 423 The development of tree-ring chronologies for the twelve most dominant tree species at our site (Helcoski 424 et al., 2019; Bourg et al., 2013) gave us the sample size to compare historical drought responses across 425 species (Fig. 3) and associated traits at a single site (see also Elliott et al., 2015). Our study reinforced 426 current understanding (see Introduction) that wood density and LMA are not reliably linked to drought 427 resistance (Table 1). Contrary to previous studies in temperate deciduous forests, we did not find an 428 association between xylem porosity and drought tolerance, as the two diffuse-porous species, Liriodendron 429 tulipifera and Faqus grandifolia, were at opposite ends of the Rt spectrum (Fig. 3). While the low Rt of L. 430 tulipifera is consistent with other studies (Elliott et al., 2015), the high Rt of F. grandifolia contrasts with 431 studies identifying diffuse porous species in general (Elliott et al., 2015; Kannenberg et al., 2019), and the 432 genus Fagus in particular (Friedrichs et al., 2009), as drought sensitive. There are two potential explanations 433 for this discrepancy. First, other traits can and do override the influence of xylem porosity on drought 434 resistance. Ring-porous species are restricted mainly to temperate deciduous forests (Wheeler et al., 2007), while highly drought-tolerant diffuse-porous species exist in other biomes (REFS). Fagus grandifolia had 436 intermediate π_{tlp} and low PLA_{dry} (Fig. S4), which would have contributed to its drought resistance (Fig. 4; 437 see discussion below). A second explanation of why F. qrandifolia trees at this particular site had higher Rt 438 is that the sampled individuals, reflective of the population within the plot, are generally shorter and in 439 less-dominant canopy positions compared to most other species (Fig. S4). The species, which is highly 440 shade-tolerant, also has deep crowns (Anderson-Teixeira et al., 2015b), implying that a lower proportion of leaves would be affected by harsher microclimatic conditions at the top of the canopy under drought (Fig. 2). 442 Thus, the high Rt of the sampled F. grandifolia population can be explained by a combination of fairly 443 drought-resistant leaf traits, shorter stature, and a buffered microenvironment. Concerted measurement of tree-rings and leaf drought tolerance traits of emerging importance (Scoffoni et al., 2014; Bartlett et al., 2016a; Medeiros et al., 2019) allowed novel insights into the role of drought 446 tolerance traits in shaping drought response. The finding that PLA_{dry} and π_{tlp} can be useful for predicting

drought responses of tree growth (Fig. 4; Table 1) is both novel and consistent with previous studies linking 448 these traits to habitat and drought tolerance. Previous studies have demonstrated that π_{tlp} and PLA_{dry} are physiologically meaningful traits linked to species distribution along moisture gradients (Maréchaux et al., 450 2015; Fletcher et al., 2018; Medeiros et al., 2019; Simeone et al., 2019; Rosas et al., 2019; Zhu et al., 2018), 451 and our findings indicate that these traits also influence drought responses. Furthermore, the observed linkage of π_{tlp} to Rt in this forest aligns with observations in the Amazon that π_{tlp} is higher in 453 drought-intolerant than drought-tolerant plant functional type. Further, it adds support to the idea that this 454 trait is useful for categorizing and representing species' drought responses in models (Powell et al., 2017). 455 Because both PLA_{dry} and π_{tlp} can be measured relatively easily (Bartlett et al., 2012; Scoffoni et al., 2014), 456 they hold promise for predicting drought growth responses across diverse forests. The importance of 457 predicting drought responses from species traits increases with tree species diversity; whereas it is feasible to study drought responses for all dominant species in most boreal and temperate forests (e.g., this study), this 459 becomes difficult to impossible for species that do not form annual rings, and for diverse tropical forests. 460 Although progress is being made for the tropics (Schöngart et al., 2017), a full linkage of drought tolerance 461 traits to drought responses would be invaluable for forecasting how little-known species and whole forests will 462 respond to future droughts (Christoffersen et al., 2016; Powell et al., 2017). 463 As climate change drives increasing drought in many of the world's forests (Trenberth et al., 2014; 464 Intergovernmental Panel on Climate Change, 2015), the fate of forests and their climate feedbacks will be shaped by the biophysical and physiological drivers observed here. Our results, consistent with other 466 observations around the world, imply that the tallest, most exposed trees will be most affected (Bennett 467 et al., 2015; Stovall et al., 2019). We show that, at least within the mature forest studied here, the vulnerability conferred by tall height and associated crown exposure outweigh any advantage of a larger root 469 system, even in drier microenvironments. This would suggest that the drought responses of trees in mature 470 forests are more strongly differentiated along the size spectrum by their above- than below-ground 471 environment. The same may not be true of systems where short trees exist outside of a buffered understory 472 environment-i.e., open grown trees or short-statured, early-successional forests. The latter appear to be 473 limited more strongly by root water access during drought (Bretfeld et al., 2018), and would also be 474 dominated by species with different traits. The earlier-successional species at our site (Liriodendron 475 tulipifera, Quercus spp., Fraxinus americana) display a mix of traits conferring drought tolerance and resistance (Table 2), while the late-successional Faqus grandifolia displayed high drought resistance, in part 477 because it exists primarily within a buffered microenvironment. Further research on how leaf drought 478 tolerance traits and drought vulnerability change over the course of succession would be valuable for 479 addressing how drought tolerance changes as forests age (e.g. Rodríguez-Catón et al., 2015). In the 480 meantime, the results of this study advance our knowledge of the factors conferring drought resistance in a 481 mature forest, opening the door for more accurate forecasting of forest responses to future drought.

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494 Author Contribution

- KAT, IM, and AJT designed the research. Tree-ring chronologies were developed by RH under guidance of
- ⁴⁹⁶ AJT and NP. Trait data were collected by IM, JZ under guidance of NK and LS. Other plot data were
- collected by IM, AS, EGA, and NB under guidance of EGA and WM. Data analyses were performed by IM
- under guidance of KAT and VH. KAT and IM interpreted the results. IM and KAT wrote the first draft of
- manuscript, and all authors contributed to revisions.

500 Supplementary Information

- Table S1: Monthly Palmer Drought Severity Index (PDSI), and its rank among all years between 1950 and
- 502 2009 (driest=1), for focal droughts.
- Table S2: Species-specific bark thickness regression equations.
- Table S3: Species-specific height regression equations.
- Table S4. Individual tests of species traits as drivers of drought resistance, where Rt is used as the response
- variable.
- Table S5. Individual tests of species traits as drivers of drought resistance, where Rt_{ARIMA} is used as the
- 508 response variable.
- Table S6. Summary of top full models for each drought instance, where Rt is used as the response variable.
- Table S7. Summary of top models for each drought instance, where Rt_{ARIMA} is used as the response
- 511 variable.
- Figure S1. Time series of Palmer Drought Severity Index (PDSI) for the 2.5 years prior to each focal drought
- 513 Figure S2: Map of ForestGEO plot showing topographic wetness index and location of cored trees
- Figure S3: Distribution of reconstructed tree heights across drought years.
- ⁵¹⁵ Figure S4. Distribution of independent variables by species.
- Figure S5. Comparison of Rt and Rt_{ARIMA} results, with residuals, for each drought scenario
- Figure S6. Visualization of best model, with data, for all droughts combined.

References

- Abrams, M. D. (1990). Adaptations and responses to drought in Quercus species of North America. *Tree Physiology*, 7(1-2-3-4):227–238.
- Allen, C. D., Breshears, D. D., and McDowell, N. G. (2015). On underestimation of global vulnerability to tree mortality and forest die-off from hotter drought in the Anthropocene. *Ecosphere*, 6(8):art129.

- Allen, C. D., Macalady, A. K., Chenchouni, H., Bachelet, D., McDowell, N., Vennetier, M., Kitzberger, T.,
- Rigling, A., Breshears, D. D., Hogg, E. H. T., Gonzalez, P., Fensham, R., Zhang, Z., Castro, J., Demidova,
- N., Lim, J.-H., Allard, G., Running, S. W., Semerci, A., and Cobb, N. (2010). A global overview of
- drought and heat-induced tree mortality reveals emerging climate change risks for forests. Forest Ecology
- $and\ Management,\ 259(4):660-684.$
- Anderegg, W. R. L., Klein, T., Bartlett, M., Sack, L., Pellegrini, A. F. A., Choat, B., and Jansen, S. (2016).
- Meta-analysis reveals that hydraulic traits explain cross-species patterns of drought-induced tree mortality
- across the globe. Proceedings of the National Academy of Sciences, 113(18):5024–5029.
- Anderson-Teixeira, K. J., Davies, S. J., Bennett, A. C., Gonzalez-Akre, E. B., Muller-Landau, H. C., Wright,
- S. J., Salim, K. A., Zambrano, A. M. A., Alonso, A., Baltzer, J. L., Basset, Y., Bourg, N. A., Broadbent,
- E. N., Brockelman, W. Y., Bunyavejchewin, S., Burslem, D. F. R. P., Butt, N., Cao, M., Cardenas, D.,
- Chuyong, G. B., Clay, K., Cordell, S., Dattaraja, H. S., Deng, X., Detto, M., Du, X., Duque, A., Erikson,
- D. L., Ewango, C. E. N., Fischer, G. A., Fletcher, C., Foster, R. B., Giardina, C. P., Gilbert, G. S.,
- Gunatilleke, N., Gunatilleke, S., Hao, Z., Hargrove, W. W., Hart, T. B., Hau, B. C. H., He, F., Hoffman,
- F. M., Howe, R. W., Hubbell, S. P., Inman-Narahari, F. M., Jansen, P. A., Jiang, M., Johnson, D. J.,
- Kanzaki, M., Kassim, A. R., Kenfack, D., Kibet, S., Kinnaird, M. F., Korte, L., Kral, K., Kumar, J.,
- Larson, A. J., Li, Y., Li, X., Liu, S., Lum, S. K. Y., Lutz, J. A., Ma, K., Maddalena, D. M., Makana, J.-R.,
- Malhi, Y., Marthews, T., Serudin, R. M., McMahon, S. M., McShea, W. J., Memiaghe, H. R., Mi, X.,
- Mizuno, T., Morecroft, M., Myers, J. A., Novotny, V., Oliveira, A. A. d., Ong, P. S., Orwig, D. A.,
- Ostertag, R., Ouden, J. d., Parker, G. G., Phillips, R. P., Sack, L., Sainge, M. N., Sang, W.,
- Sri-ngernyuang, K., Sukumar, R., Sun, I.-F., Sungpalee, W., Suresh, H. S., Tan, S., Thomas, S. C.,
- Thomas, D. W., Thompson, J., Turner, B. L., Uriarte, M., Valencia, R., Vallejo, M. I., Vicentini, A., Vrška,
- T., Wang, X., Wang, X., Weiblen, G., Wolf, A., Xu, H., Yap, S., and Zimmerman, J. (2015a).
- 546 CTFS-ForestGEO: a worldwide network monitoring forests in an era of global change. Global Change
- 547 Biology, 21(2):528-549.
- Anderson-Teixeira, K. J., McGarvey, J. C., Muller-Landau, H. C., Park, J. Y., Gonzalez-Akre, E. B.,
- Herrmann, V., Bennett, A. C., So, C. V., Bourg, N. A., Thompson, J. R., McMahon, S. M., and McShea,
- W. J. (2015b). Size-related scaling of tree form and function in a mixed-age forest. Functional Ecology,
- ⁵⁵¹ 29(12):1587–1602.
- 552 Bartlett, M. K., Klein, T., Jansen, S., Choat, B., and Sack, L. (2016a). The correlations and sequence of
- plant stomatal, hydraulic, and wilting responses to drought. Proceedings of the National Academy of
- Sciences, 113(46):13098-13103.
- Bartlett, M. K., Scoffoni, C., Ardy, R., Zhang, Y., Sun, S., Cao, K., and Sack, L. (2012). Rapid
- determination of comparative drought tolerance traits: using an osmometer to predict turgor loss point.
- Methods in Ecology and Evolution, 3(5):880–888.
- Bartlett, M. K., Zhang, Y., Yang, J., Kreidler, N., Sun, S.-W., Lin, L., Hu, Y.-H., Cao, K.-F., and Sack, L.
- 559 (2016b). Drought tolerance as a driver of tropical forest assembly: resolving spatial signatures for multiple
- processes. Ecology, 97(2):503-514.
- Bates, D., Maechler, M., Bolker, B., and Walker, S. (2019). *lme4: Linear Mixed-Effects Models using 'Eigen'*and S4. R package version 1.1-21.

- Bennett, A. C., McDowell, N. G., Allen, C. D., and Anderson-Teixeira, K. J. (2015). Larger trees suffer most 563 during drought in forests worldwide. Nature Plants, 1(10):15139.
- Beven, K. J. and Kirkby, M. J. (1979). A physically based, variable contributing area model of basin
- hydrology / Un modèle à base physique de zone d'appel variable de l'hydrologie du bassin versant. 566
- Hydrological Sciences Bulletin, 24(1):43–69. 567
- Bonan, G. B. (2008). Forests and Climate Change: Forcings, Feedbacks, and the Climate Benefits of Forests. 568 Science, 320(5882):1444-1449.
- Bourg, N. A., McShea, W. J., Thompson, J. R., McGarvey, J. C., and Shen, X. (2013). Initial census, woody 570 seedling, seed rain, and stand structure data for the SCBI SIGEO Large Forest Dynamics Plot. Ecology, 571
- 94(9):2111-2112. 572
- Bretfeld, M., Ewers, B. E., and Hall, J. S. (2018). Plant water use responses along secondary forest 573
- succession during the 2015–2016 El Niño drought in Panama. New Phytologist, 219(3):885–899. 574
- Brewer, M. J., Butler, A., and Cooksley, S. L. (2016). The relative performance of AIC, AICC and BIC in 575 the presence of unobserved heterogeneity. Methods in Ecology and Evolution, 7(6):679-692. 576
- Campbell, G. S. and Norman, J. M. (1998). An Introduction to Environmental Biophysics, volume 2nd. 577 Springer, New York. 578
- Christoffersen, B. O., Gloor, M., Fauset, S., Fyllas, N. M., Galbraith, D. R., Baker, T. R., Kruijt, B., 579
- Rowland, L., Fisher, R. A., Binks, O. J., Sevanto, S., Xu, C., Jansen, S., Choat, B., Mencuccini, M., 580
- McDowell, N. G., and Meir, P. (2016). Linking hydraulic traits to tropical forest function in a 581
- size-structured and trait-driven model (TFS v.1-Hydro). 582
- Clark, J. S., Iverson, L., Woodall, C. W., Allen, C. D., Bell, D. M., Bragg, D. C., D'Amato, A. W., Davis, 583
- F. W., Hersh, M. H., Ibanez, I., Jackson, S. T., Matthews, S., Pederson, N., Peters, M., Schwartz, M. W., 584
- Waring, K. M., and Zimmermann, N. E. (2016). The impacts of increasing drought on forest dynamics, 585
- structure, and biodiversity in the United States. Global Change Biology, 22(7):2329–2352. 586
- Condit, R. (1998). Tropical Forest Census Plots: Methods and Results from Barro Colorado Island, Panama 587 and a Comparison with Other Plots. Springer Berlin Heidelberg, Berlin, Heidelberg. 588
- Cook, B. I., Ault, T. R., and Smerdon, J. E. (2015). Unprecedented 21st century drought risk in the 589 American Southwest and Central Plains. Science Advances, 1(1):e1400082. 590
- Couvreur, V., Ledder, G., Manzoni, S., Way, D. A., Muller, E. B., and Russo, S. E. (2018). Water transport 591 through tall trees: A vertically explicit, analytical model of xylem hydraulic conductance in stems. Plant,
- Cell & Environment, 41(8):1821–1839. 593

592

- Dai, A., Zhao, T., and Chen, J. (2018). Climate Change and Drought: a Precipitation and Evaporation Perspective. Current Climate Change Reports, 4(3):301–312. 595
- Davis, K. T., Dobrowski, S. Z., Holden, Z. A., Higuera, P. E., and Abatzoglou, J. T. (2019). Microclimatic 596 buffering in forests of the future: the role of local water balance. Ecography, 42(1):1-11. 597
- DeSoto, L., Cailleret, M., Sterck, F., Jansen, S., Kramer, K., Robert, E. M. R., Aakala, T., Amoroso, M. M., 598
- Bigler, C., Camarero, J. J., Čufar, K., Gea-Izquierdo, G., Gillner, S., Haavik, L. J., Heres, A.-M., Kane, 599
- J. M., Kharuk, V. I., Kitzberger, T., Klein, T., Levanič, T., Linares, J. C., Mäkinen, H., Oberhuber, W., 600

- Papadopoulos, A., Rohner, B., Sangüesa-Barreda, G., Stojanovic, D. B., Suárez, M. L., Villalba, R., and
- Martínez-Vilalta, J. (2020). Low growth resilience to drought is related to future mortality risk in trees.
- Nature Communications, 11(1):545.
- 604 Detto, M., Muller-Landau, H. C., Mascaro, J., and Asner, G. P. (2013). Hydrological Networks and
- Associated Topographic Variation as Templates for the Spatial Organization of Tropical Forest Vegetation.
- PLOS ONE, 8(10):e76296.
- Druckenbrod, D. L., Martin-Benito, D., Orwig, D. A., Pederson, N., Poulter, B., Renwick, K. M., and
- Shugart, H. H. (2019). Redefining temperate forest responses to climate and disturbance in the eastern
- United States: New insights at the mesoscale. Global Ecology and Biogeography, 28(5):557–575.
- Elliott, K. J., Miniat, C. F., Pederson, N., and Laseter, S. H. (2015). Forest tree growth response to
- hydroclimate variability in the southern Appalachians. Global Change Biology, 21(12):4627–4641.
- Enquist, B. J. and Niklas, K. J. (2002). Global Allocation Rules for Patterns of Biomass Partitioning in Seed
 Plants. Science, 295(5559):1517–1520.
- Farrell, C., Szota, C., and Arndt, S. K. (2017). Does the turgor loss point characterize drought response in dryland plants? *Plant, Cell & Environment*, 40(8):1500–1511.
- Fletcher, L. R., Cui, H., Callahan, H., Scoffoni, C., John, G. P., Bartlett, M. K., Burge, D. O., and Sack, L.
- (2018). Evolution of leaf structure and drought tolerance in species of Californian Ceanothus. American
- Journal of Botany, 105(10):1672–1687.
- ⁶¹⁹ Friedlingstein, P., Cox, P., Betts, R., Bopp, L., von Bloh, W., Brovkin, V., Cadule, P., Doney, S., Eby, M.,
- Fung, I., Bala, G., John, J., Jones, C., Joos, F., Kato, T., Kawamiya, M., Knorr, W., Lindsay, K.,
- Matthews, H. D., Raddatz, T., Rayner, P., Reick, C., Roeckner, E., Schnitzler, K.-G., Schnur, R.,
- Strassmann, K., Weaver, A. J., Yoshikawa, C., and Zeng, N. (2006). Climate-Carbon Cycle Feedback
- Analysis: Results from the C4MIP Model Intercomparison. Journal of Climate, 19(14):3337–3353.
- ⁶²⁴ Friedrichs, D. A., Trouet, V., Büntgen, U., Frank, D. C., Esper, J., Neuwirth, B., and Löffler, J. (2009).
- Species-specific climate sensitivity of tree growth in Central-West Germany. Trees, 23(4):729.
- ⁶²⁶ Gillerot, L., Forrester, D. I., Bottero, A., Rigling, A., and Lévesque, M. (2020). Tree Neighbourhood
- Diversity Has Negligible Effects on Drought Resilience of European Beech, Silver Fir and Norway Spruce.
- Ecosystems.
- 629 Gonzalez-Akre, E., Meakem, V., Eng, C.-Y., Tepley, A. J., Bourg, N. A., McShea, W., Davies, S. J., and
- Anderson-Teixeira, K. (2016). Patterns of tree mortality in a temperate deciduous forest derived from a
- large forest dynamics plot. *Ecosphere*, 7(12):e01595.
- Greenwood, S., Ruiz-Benito, P., Martínez-Vilalta, J., Lloret, F., Kitzberger, T., Allen, C. D., Fensham, R.,
- Laughlin, D. C., Kattge, J., Bönisch, G., Kraft, N. J. B., and Jump, A. S. (2017). Tree mortality across
- biomes is promoted by drought intensity, lower wood density and higher specific leaf area. *Ecology Letters*,
- 20(4):539-553.
- 656 Guerfel, M., Baccouri, O., Boujnah, D., Chaïbi, W., and Zarrouk, M. (2009). Impacts of water stress on gas
- exchange, water relations, chlorophyll content and leaf structure in the two main Tunisian olive (Olea
- europaea L.) cultivars. Scientia Horticulturae, 119(3):257–263.

- Hacket-Pain, A. J., Cavin, L., Friend, A. D., and Jump, A. S. (2016). Consistent limitation of growth by
 high temperature and low precipitation from range core to southern edge of European beech indicates
 widespread vulnerability to changing climate. European Journal of Forest Research, 135(5):897–909.
- Harris, I., Jones, P. D., Osborn, T. J., and Lister, D. H. (2014). Updated high-resolution grids of monthly climatic observations the CRU TS3.10 Dataset. *International Journal of Climatology*, 34(3):623–642.
- Helcoski, R., Tepley, A. J., Pederson, N., McGarvey, J. C., Meakem, V., Herrmann, V., Thompson, J. R.,
 and Anderson-Teixeira, K. J. (2019). Growing season moisture drives interannual variation in woody
 productivity of a temperate deciduous forest. New Phytologist, 0(0).
- Hoffmann, W. A., Marchin, R. M., Abit, P., and Lau, O. L. (2011). Hydraulic failure and tree dieback are
 associated with high wood density in a temperate forest under extreme drought. Global Change Biology,
 17(8):2731–2742.
- Hollister, J. (2018). elevatr: Access Elevation Data from Various APIs. R package version 0.2.0.
- Hyndman, R., Athanasopoulos, G., Bergmeir, C., Caceres, G., Chhay, L., O'Hara-Wild, M., Petropoulos, F.,
 Razbash, S., Wang, E., and Yasmeen, F. (2020). forecast: Forecasting Functions for Time Series and
 Linear Models. R package version 8.12.
- Intergovernmental Panel on Climate Change (2015). Climate Change 2014: Impacts, Adaptation and
 Vulnerability: Working Group II Contribution to the IPCC Fifth Assessment Report. Volume 2 Volume 2.
 OCLC: 900892773.
- Jennings, S. B., Brown, N. D., and Sheil, D. (1999). Assessing forest canopies and understorey illumination: canopy closure, canopy cover and other measures. *Forestry: An International Journal of Forest Research*, 72(1):59–74.
- Kannenberg, S. A., Novick, K. A., Alexander, M. R., Maxwell, J. T., Moore, D. J. P., Phillips, R. P., and
 Anderegg, W. R. L. (2019). Linking drought legacy effects across scales: From leaves to tree rings to
 ecosystems. Global Change Biology, 0(ja).
- 663 Katabuchi, M. (2019). LeafArea: Rapid Digital Image Analysis of Leaf Area. R package version 0.1.8.
- Kennedy, D., Swenson, S., Oleson, K. W., Lawrence, D. M., Fisher, R., Costa, A. C. L. d., and Gentine, P.
 (2019). Implementing Plant Hydraulics in the Community Land Model, Version 5. *Journal of Advances in Modeling Earth Systems*, 11(2):485–513.
- Koike, T., Kitao, M., Maruyama, Y., Mori, S., and Lei, T. T. (2001). Leaf morphology and photosynthetic
 adjustments among deciduous broad-leaved trees within the vertical canopy profile. *Tree Physiology*,
 21(12-13):951-958.
- Kunert, N., Aparecido, L. M. T., Wolff, S., Higuchi, N., Santos, J. d., Araujo, A. C. d., and Trumbore, S.
 (2017). A revised hydrological model for the Central Amazon: The importance of emergent canopy trees in
 the forest water budget. Agricultural and Forest Meteorology, 239:47–57.
- Larjavaara, M. and Muller-Landau, H. C. (2013). Measuring tree height: a quantitative comparison of two common field methods in a moist tropical forest. *Methods in Ecology and Evolution*, 4(9):793–801.
- Liu, H., Gleason, S. M., Hao, G., Hua, L., He, P., Goldstein, G., and Ye, Q. (2019). Hydraulic traits are coordinated with maximum plant height at the global scale. *Science Advances*, 5(2):eaav1332.

- Liu, Y. and Muller, R. N. (1993). Effect of Drought and Frost on Radial Growth of Overstory and Understory Stems in a Deciduous Forest. The American Midland Naturalist, 129(1):19-25.
- Lloret, F., Keeling, E. G., and Sala, A. (2011). Components of tree resilience: effects of successive low-growth episodes in old ponderosa pine forests. Oikos, 120(12):1909–1920. 680
- Lunch, C., Laney, C., Mietkiewicz, N., Sokol, E., Cawley, K., and NEON (National Ecological Observatory 681 Network) (2020). neon Utilities: Utilities for Working with NEON Data. R package version 1.3.5. 682
- Martin-Benito, D. and Pederson, N. (2015). Convergence in drought stress, but a divergence of climatic 683 drivers across a latitudinal gradient in a temperate broadleaf forest. Journal of Biogeography, 684 42(5):925-937. 685
- Maréchaux, I., Bartlett, M. K., Sack, L., Baraloto, C., Engel, J., Joetzjer, E., and Chave, J. (2015). Drought 686 tolerance as predicted by leaf water potential at turgor loss point varies strongly across species within an 687 Amazonian forest. Functional Ecology, 29(10):1268–1277. 688
- Maréchaux, I., Saint-André, L., Bartlett, M. K., Sack, L., and Chave, J. (2019). Leaf drought tolerance 689 cannot be inferred from classic leaf traits in a tropical rainforest. Journal of Ecology. 690
- Mazerolle, M. J. and portions of code contributed by Dan Linden. (2019). AICcmodavg: Model Selection and 691 Multimodel Inference Based on (Q)AIC(c). R package version 2.2-2. 692
- McDowell, N. G. and Allen, C. D. (2015). Darcy's law predicts widespread forest mortality under climate 693 warming. Nature Climate Change, 5(7):669-672. 694
- McDowell, N. G., Bond, B. J., Dickman, L. T., Ryan, M. G., and Whitehead, D. (2011). Relationships 695
- Between Tree Height and Carbon Isotope Discrimination. In Meinzer, F. C., Lachenbruch, B., and
- Dawson, T. E., editors, Size- and Age-Related Changes in Tree Structure and Function, Tree Physiology, 697
- pages 255–286. Springer Netherlands, Dordrecht. 698
- Meakem, V., Tepley, A. J., Gonzalez-Akre, E. B., Herrmann, V., Muller-Landau, H. C., Wright, S. J., 699
- Hubbell, S. P., Condit, R., and Anderson-Teixeira, K. J. (2018). Role of tree size in moist tropical forest 700 carbon cycling and water deficit responses. New Phytologist, 219(3):947–958. 701
- Medeiros, C. D., Scoffoni, C., John, G. P., Bartlett, M. K., Inman-Narahari, F., Ostertag, R., Cordell, S., 702 Giardina, C., and Sack, L. (2019). An extensive suite of functional traits distinguishes Hawaiian wet and 703 dry forests and enables prediction of species vital rates. Functional Ecology, 33(4):712–734.
- Meinzer, F. C., Andrade, J. L., Goldstein, G., Holbrook, N. M., Cavelier, J., and Wright, S. J. (1999). 705
- Partitioning of soil water among canopy trees in a seasonally dry tropical forest. Oecologia, 121(3):293-301. 706
- Mencuccini, M. (2003). The ecological significance of long-distance water transport: short-term regulation, 707
- long-term acclimation and the hydraulic costs of stature across plant life forms. Plant, Cell & 708
- Environment, 26(1):163-182. 709

704

- Metcalfe, P., Beven, K., and Freer, J. (2018). dynatopmodel: Implementation of the Dynamic TOPMODEL 710 Hydrological Model. R package version 1.2.1. 711
- NEON (2018). National Ecological Observatory Network. 2016, 2017, 2018. Data Products: DP1.00001.001,
- DP1.00098.001, DP1.00002.001. Provisional data downloaded from http://data.neonscience.org/ in May 713
- 2019. Battelle, Boulder, CO, USA. 714

- Olson, M., Rosell, J. A., Martínez-Pérez, C., León-Gómez, C., Fajardo, A., Isnard, S., Cervantes-Alcayde,
- M. A., Echeverría, A., Figueroa-Abundiz, V. A., Segovia-Rivas, A., Trueba, S., and Vázquez-Segovia, K.
- 717 (2020). Xylem vessel-diameter—shoot-length scaling: ecological significance of porosity types and other
- traits. Ecological Monographs, n/a(n/a).
- Olson, M. E., Anfodillo, T., Rosell, J. A., Petit, G., Crivellaro, A., Isnard, S., León-Gómez, C.,
- Alvarado-Cárdenas, L. O., and Castorena, M. (2014). Universal hydraulics of the flowering plants: vessel
- diameter scales with stem length across angiosperm lineages, habits and climates. Ecology Letters,
- ⁷²² 17(8):988–997.
- Olson, M. E., Soriano, D., Rosell, J. A., Anfodillo, T., Donoghue, M. J., Edwards, E. J., León-Gómez, C.,
- Dawson, T., Martínez, J. J. C., Castorena, M., Echeverría, A., Espinosa, C. I., Fajardo, A., Gazol, A.,
- Isnard, S., Lima, R. S., Marcati, C. R., and Méndez-Alonzo, R. (2018). Plant height and hydraulic
- vulnerability to drought and cold. Proceedings of the National Academy of Sciences, 115(29):7551–7556.
- Phillips, N. G., Ryan, M. G., Bond, B. J., McDowell, N. G., Hinckley, T. M., and Čermák, J. (2003).
- Reliance on stored water increases with tree size in three species in the Pacific Northwest. Tree Physiology,
- 23(4):237-245.
- Powell, T. L., Wheeler, J. K., Oliveira, A. A. R. d., Costa, A. C. L. d., Saleska, S. R., Meir, P., and
- Moorcroft, P. R. (2017). Differences in xylem and leaf hydraulic traits explain differences in drought
- tolerance among mature Amazon rainforest trees. Global Change Biology, 23(10):4280–4293.
- Pretzsch, H., Schütze, G., and Biber, P. (2018). Drought can favour the growth of small in relation to tall
- trees in mature stands of Norway spruce and European beech. Forest Ecosystems, 5(1):20.
- R Core Team (2019). R: A Language and Environment for Statistical Computing. R Foundation for
- 736 Statistical Computing, Vienna, Austria.
- Rey-Sánchez, A. C., Slot, M., Posada, J. M., and Kitajima, K. (2016). Spatial and seasonal variation in leaf
- temperature within the canopy of a tropical forest. Climate Research, 71(1):75-89.
- 739 Rodríguez-Catón, M., Villalba, R., Srur, A. M., and Luckman, B. (2015). Long-term trends in radial growth
- associated with Nothofagus pumilio forest decline in Patagonia: Integrating local- into regional-scale
- patterns. Forest Ecology and Management, 339:44–56.
- Rosas, T., Mencuccini, M., Barba, J., Cochard, H., Saura-Mas, S., and Martínez-Vilalta, J. (2019).
- Adjustments and coordination of hydraulic, leaf and stem traits along a water availability gradient. New
- 744 Phytologist, 223(2):632–646.
- Roskilly, B., Keeling, E., Hood, S., Giuggiola, A., and Sala, A. (2019). Conflicting functional effects of xylem
- pit structure relate to the growth-longevity trade-off in a conifer species. PNAS. doi:
- 747 /10.1073/pnas.1900734116.
- Ryan, M. G., Phillips, N., and Bond, B. J. (2006). The hydraulic limitation hypothesis revisited. *Plant, Cell*
- ⁷⁴⁹ & Environment, 29(3):367–381.
- ⁷⁵⁰ Sapes, G., Roskilly, B., Dobrowski, S., Maneta, M., Anderegg, W. R. L., Martinez-Vilalta, J., and Sala, A.
- ₇₅₁ (2019). Plant water content integrates hydraulics and carbon depletion to predict drought-induced seedling
- mortality. *Tree Physiology*, 39(8):1300–1312.

- Scharnweber, T., Heinze, L., Cruz-García, R., van der Maaten-Theunissen, M., and Wilmking, M. (2019).
- Confessions of solitary oaks: We grow fast but we fear the drought. Dendrochronologia, 55:43–49.
- 755 Scholz, F. G., Phillips, N. G., Bucci, S. J., Meinzer, F. C., and Goldstein, G. (2011). Hydraulic Capacitance:
- ₇₅₆ Biophysics and Functional Significance of Internal Water Sources in Relation to Tree Size. In Meinzer,
- F. C., Lachenbruch, B., and Dawson, T. E., editors, Size- and Age-Related Changes in Tree Structure and
- Function, Tree Physiology, pages 341–361. Springer Netherlands, Dordrecht.
- Schöngart, J., Bräuning, A., Barbosa, A. C. M. C., Lisi, C. S., and de Oliveira, J. M. (2017).
- Dendroecological Studies in the Neotropics: History, Status and Future Challenges. In Amoroso, M. M.,
- Daniels, L. D., Baker, P. J., and Camarero, J. J., editors, Dendroecology: Tree-Ring Analyses Applied to
- ⁷⁶² Ecological Studies, Ecological Studies, pages 35–73. Springer International Publishing, Cham.
- Scoffoni, C., Vuong, C., Diep, S., Cochard, H., and Sack, L. (2014). Leaf Shrinkage with Dehydration:
- Coordination with Hydraulic Vulnerability and Drought Tolerance. Plant Physiology, 164(4):1772–1788.
- ⁷⁶⁵ Simeone, C., Maneta, M. P., Holden, Z. A., Sapes, G., Sala, A., and Dobrowski, S. Z. (2019). Coupled
- ecohydrology and plant hydraulics modeling predicts ponderosa pine seedling mortality and lower treeline
- in the US Northern Rocky Mountains. New Phytologist, 221(4):1814–1830.
- 768 Slette, I. J., Post, A. K., Awad, M., Even, T., Punzalan, A., Williams, S., Smith, M. D., and Knapp, A. K.
- (2019). How ecologists define drought, and why we should do better. Global Change Biology, 0(0):1–8.
- Stovall, A. E. L., Anderson-Teixeira, K. J., and Shugart, H. H. (2018a). Assessing terrestrial laser scanning
- for developing non-destructive biomass allometry. Forest Ecology and Management, 427:217–229.
- 772 Stovall, A. E. L., Anderson-Teixeira, K. J., and Shugart, H. H. (2018b). Terrestrial LiDAR-derived
- non-destructive woody biomass estimates for 10 hardwood species in Virginia. Data in Brief, 19:1560–1569.
- 5774 Stovall, A. E. L., Shugart, H., and Yang, X. (2019). Tree height explains mortality risk during an intense
- drought. Nature Communications, 10(1):1–6.
- 576 Stovall, A. E. L., Shugart, H. H., and Yang, X. (2020). Reply to "Height-related changes in forest
- composition explain increasing tree mortality with height during an extreme drought". Nature
- Communications, 11(1):3401.
- 779 Suarez, M. L., Ghermandi, L., and Kitzberger, T. (2004). Factors predisposing episodic drought-induced tree
- mortality in Nothofagus—site, climatic sensitivity and growth trends. Journal of Ecology, 92(6):954–966.
- 781 Sørensen, R., Zinko, U., and Seibert, J. (2006). On the calculation of the topographic wetness index:
- evaluation of different methods based on field observations. Hydrology and Earth System Sciences,
- 783 10(1):101–112.
- Trenberth, K. E., Dai, A., van der Schrier, G., Jones, P. D., Barichivich, J., Briffa, K. R., and Sheffield, J.
- 785 (2014). Global warming and changes in drought. Nature Climate Change, 4(1):17–22.
- Wheeler, E. A., Baas, P., and Rodgers, S. (2007). Variations In Dieot Wood Anatomy: A Global Analysis
- Based on the Insidewood Database. IAWA Journal, 28(3):229–258.
- ⁷⁸⁸ Zach, A., Schuldt, B., Brix, S., Horna, V., Culmsee, H., and Leuschner, C. (2010). Vessel diameter and xylem
- hydraulic conductivity increase with tree height in tropical rainforest trees in Sulawesi, Indonesia. Flora-
- Morphology, Distribution, Functional Ecology of Plants, 205(8):506–512.

- Zellweger, F., Coomes, D., Lenoir, J., Depauw, L., Maes, S. L., Wulf, M., Kirby, K. J., Brunet, J., Kopecký,
- M., Máliš, F., Schmidt, W., Heinrichs, S., Ouden, J. d., Jaroszewicz, B., Buyse, G., Spicher, F., Verheyen,
- K., and Frenne, P. D. (2019). Seasonal drivers of understorey temperature buffering in temperate
- deciduous forests across Europe. Global Ecology and Biogeography, 28(12):1774–1786.
- ⁷⁹⁵ Zhu, S.-D., Chen, Y.-J., Ye, Q., He, P.-C., Liu, H., Li, R.-H., Fu, P.-L., Jiang, G.-F., and Cao, K.-F. (2018).
- Leaf turgor loss point is correlated with drought tolerance and leaf carbon economics traits. Tree
- 797 Physiology, 38(5):658–663.
- ⁷⁹⁸ Zuleta, D., Duque, A., Cardenas, D., Muller-Landau, H. C., and Davies, S. J. (2017). Drought-induced
- mortality patterns and rapid biomass recovery in a terra firme forest in the Colombian Amazon. *Ecology*,
- 98(10):2538-2546.